



# Investigating three-dimensional localised effects of age, disease and treatment on mouse bone geometry using Principal Component Analysis

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## Osteoporosis:

The prevalence and severity of bone breakages and fractures drastically increases with age. Osteoporosis leads to increased possibility of fragility fractures.

## Pre-Clinical studies:

- Murine bone models are widely used.
- Osteoporosis and treatment strategies are simulated.
- Exercise is simulated by applying in-vivo mechanical loading of long bones.
- Treatment impact on bone geometry is typically assessed as changes in averaged scalar morphometric parameters.
- However, the precision and accuracy of these studies are limited by approximations - cylindrical models.

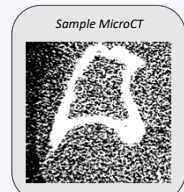
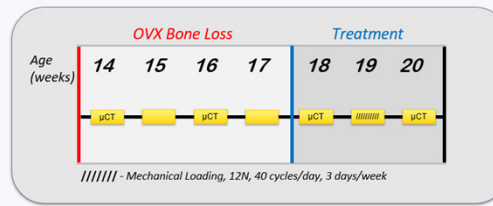
## Study Aims:

- Use Principal Component Analysis to analyze variations in mice tibia geometry.

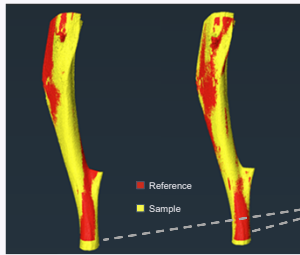
## 0) Data

### Data Collection:

- Mice are ovariectomized at week 14 to induce menopausal osteoporosis.
- Mice are subjected to mechanical loading at week 19.
- Right tibia is scanned through MicroCT.
- Longitudinal data of 6 treated mice, 2 ages.



## 1) Pre-Processing



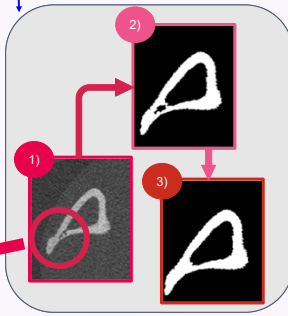
### Sample registration

Performs a **rigid registration** of each sample to the reference bone, resulting in an **alignment of all samples**.



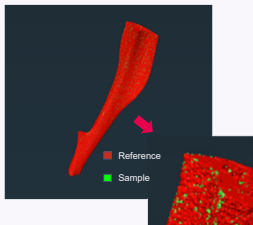
### Binarization & Feature Removal

The **images are binarized** and features such as cortical pore and trabeculae are removed to allow **topological similarity** across samples.



### Elastic Registration using SHIRT

- The reference image is mapped to all other image samples using elastic registration.
- The surface mesh of the reference is extracted from the binarized image.
- The **deformation field** is applied on the reference surface mesh.



## 2) PCA

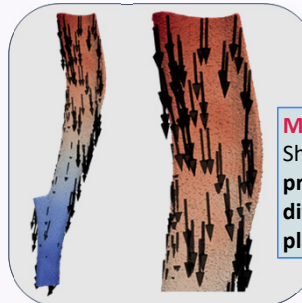
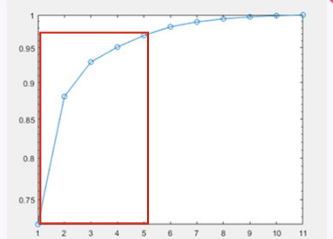
Decomposes variations into **11 orthogonal modes**, each mode portraying a unique feature. Modes are **ordered by latent** from eigen values. Variations due to **age, growth and disease** are given by PCA modes.

## 3) Results & Conclusion

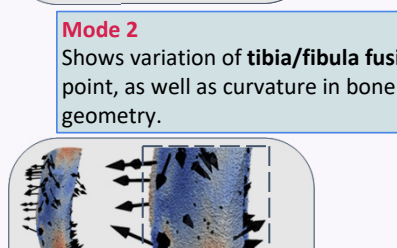
### Analysis

The **first 5 modes describe 97% of total variation**.

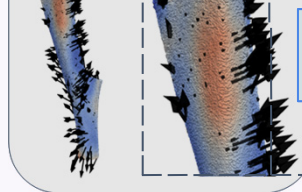
Focus is on the **first 5 modes**, as their variations are significant.



**Mode 1**  
Shows stronger variation toward **proximal end** of tibia. Caused by differences in **growth plate placement**.

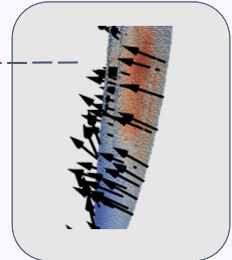


**Mode 2**  
Shows variation of **tibia/fibula fusion point**, as well as curvature in bone geometry.



**Mode 4**  
Shows variations at the **medial aspect of proximal diaphysis**.

X axis slice



### Conclusion

PCA modes show many variations, some artefacts of methodology, however some, significantly, are not. Variations may be linked to treatment – further investigation.

## In partnership with:

- Doncaster and Bassetlaw Teaching Hospitals NHS Foundation Trust
- Sheffield Teaching Hospitals NHS Foundation Trust
- Sheffield Children's NHS Foundation Trust